The Fire Model Intercomparison Project (FireMIP)

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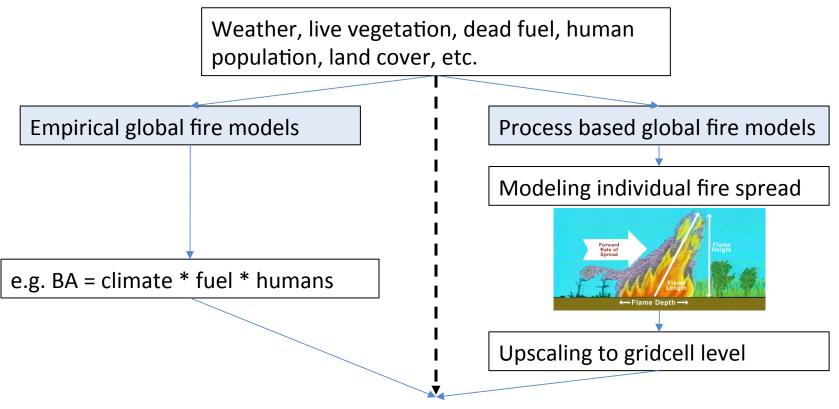


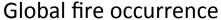


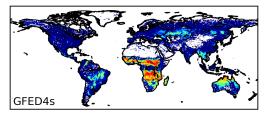




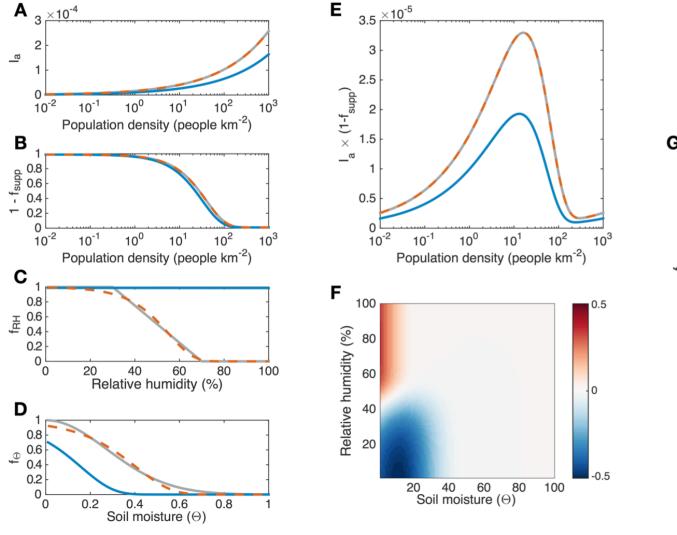
Global fire models

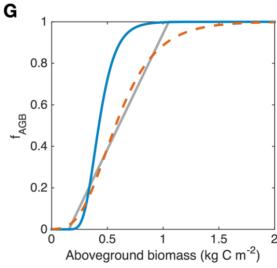






Process-based models





Rabin et al. (2018)

Global fire models over the years



The Fire Model Intercomparison Project is born

- Led from KIT by Stijn Hantson (now UC Irvine) and Prof. Almut Arneth
- First workshop September 2014

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The status and challenge of global fire modelling

Stijn Hantson¹, Almut Arneth¹, Sandy P. Harrison^{2,3}, Douglas I. Kelley^{2,3}, I. Colin Prentice^{3,4}, Sam S. Rabin⁵, Sally Archibald^{6,7}, Florent Mouillot⁸, Steve R. Arnold⁹, Paulo Artaxo¹⁰, Dominique Bachelet^{11,12}, Philippe Ciais¹³, Matthew Forrest¹⁴, Pierre Friedlingstein¹⁵, Thomas Hickler^{14,16}, Jed O. Kaplan¹⁷, Silvia Kloster¹⁸, Wolfgang Knorr¹⁹, Gitta Lasslop¹⁸, Fang Li²⁰, Stephane Mangeon²¹, Joe R. Melton²², Andrea Meyn²³, Stephen Sitch²⁴, Allan Spessa^{25,26}, Guido R. van der Werf²⁷, Apostolos Voulgarakis²¹, and Chao Yue¹³

FireMIP goals & expected outcomes

- Test the ability of models to reproduce current patterns and historical trends of fire occurrence and impacts
- Compare models' performance to identify which model structures seem particularly useful, and where more development should be focused
- Generate estimates (with uncertainty) of fire patterns and trends in the deep past (LGM), 20th century, and 21st century

FireMIP principles

- Unified input data for climate, CO₂, population, and lightning
- Observational data for evaluation should be independent of any models
- Benchmarking after Kelley et al. (2013) for standardized comparison metrics

FireMIP phase 1

- Focus on 1901 to present
- Transient 1700-2013 runs with time-evolving input data.
- Sensitivity runs:
 - World without fire.
 - Pre-industrial fire regime.
 - *Pre-industrial atmospheric CO*₂.
 - Fixed human population density.
 - Fixed 1901-1920 lightning.
 - Fixed 1901-1920 climate.
 - Fixed 1700 land use.

Model structure comparisons

Table 2. List of models participating in FireMIP, including contact person's email and key references. Also included is information relating to the configuration to be used in this phase of FireMIP. Note that "Resolution" refers to spatial and temporal resolution of the fire model only; the associated land or vegetation may update more frequently.

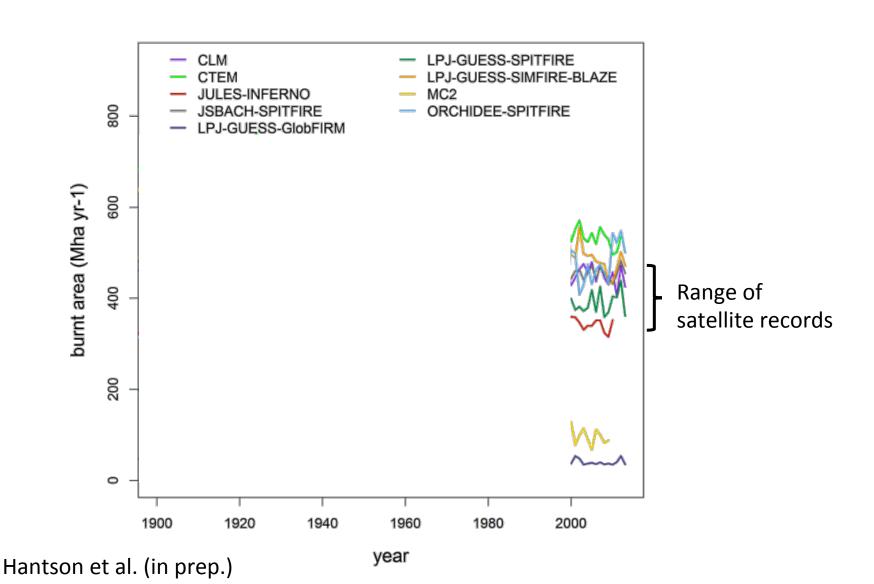
Fire model	Land/vegetation model	Dynamic vegetation		N cycle?	No. PFTs	No. soil layers	No. classes	litter	Resolution	Contact	
		Physiology	LAI, biomass	Biogeography	/						
CLM-Li fire module (Li	CLM4.5-BGC (Oleson	Yes	Yes	Yes, but not	Yes	17	15	1		~1.9° lat.	Fang Li (lifang@mail.iap.ac.cn)

Table S13. Calculation of fractional combustion for models that classify dead fuels by size. Note that differences among SPITFIRE-based models in combustion of live fuels arose as a result of inconsistencies in Thonicke et al. (2010). PFT-specific values (Tables S20, S25–S28) are denoted with a "hat." "n/a" indicates models that do not include the indicated variable or concept.

Component	JSBACH-SPITFIRE	LPJ-GUESS-SPITFIRE	LPJ-LMfire	MC-Fire	ORCHIDEE-SPITFIRE		
$FC_{d,1h}$	$G_{ag} \times (1 - \mu_S) \times \begin{cases} 1.0 & \omega_{ff/\star} < 0.18 \\ X - 0.62 \omega_{ff/\star} & 0.18 \leq \omega_{ff/\star} \leq 0.73, \\ 2.45 - 2.45 \omega_{ff/\star} & \omega_{ff/\star} > 0.73 \end{cases}$ where $X = 1.11$.	As JSBACH-SPITFIRE, but using $\omega_{d,1h/*}$ instead of $\omega_{ff/*}$, and without multiplying by $(1-\mu_S)$.	As JSBACH-SPITFIRE, but with $X=1.10$ and using $\frac{\omega_{d,1h}}{0.404}$ instead $\omega_{ff/*}$.	$0.9 imes G_{ag}$ of	$G_{ag} \times \begin{cases} \min \left(0.9, 1.11 - 0.62 \omega_{ff/*} \right) & \omega_{ff/*} \leq 0.73 \\ 2.45 - 2.45 \omega_{ff/*} & \omega_{ff/*} > 0.73 \end{cases}$		
$FC_{d,10h}$	$G_{ag} \times (1 - \mu_S) \times \begin{cases} 1.0 & \omega_{o/\star} < Y \\ 1.09 - 0.72\omega_{o/\star} & Y \leq \omega_{o/\star} \leq 0.51, \\ 1.47 - 1.47\omega_{o/\star} & \omega_{o/\star} > 0.51 \end{cases}$ where $Y = 0.13$.	As JSBACH-SPITFIRE, but with $Y=0.12$ and without multiplying by $(1-\mu_S)$.	As JSBACH-SPITFIRE, but with $Y=0.12$ and using $\frac{\omega_{d,10h}}{0.487}$ instead $\omega_{o/*}$.	$\begin{array}{c} -0.048132 + (0.917393 \times L_{d,10h} \times Z) \\ \text{I of} & \frac{L_{d,10h} \times Z}{\text{version factor } Z = 0.5 \times 0.224170} \text{ converts from} \\ \text{kg C m}^{-2} \text{ to T DM ac}^{-1}. \end{array}$	$G_{ag} \times \begin{cases} \min \left(0.9, 1.09 - 0.72 \omega_{o/\star} \right) & \omega_{o/\star} \leq 0.51 \\ 1.47 - 1.47 \omega_{o/\star} & \omega_{o/\star} > 0.51 \end{cases}$		
$FC_{d,100h}$	$G_{ag} \times (1 - \mu_S) \times min(0.45, [0.98 - 0.85 \times \omega_{d,100h}])$	$\begin{cases} 0.98 - 0.85\omega_{o/\star} & \omega_{o/\star} \leq 0.38 \\ 1.06 - 1.06\omega_{o/\star} & \omega_{o/\star} > 0.38 \end{cases}$	$\begin{cases} (1 - \mu_S) \times \\ 0.98 - 0.85 \frac{\omega_{d,100h}}{0.525} & \frac{\omega_{d,100h}}{0.525} \\ 1.06 - 1.06 \frac{\omega_{d,100h}}{0.525} & \frac{\omega_{d,100h}}{0.525} > \end{cases}$	$ \begin{array}{c} (-0.124649 + [0.869309 \times L_{d,100h} \times Z] - [48.04 \times \\ 0.38 \omega_{duff}]) \times (L_{d,100h} \times Z)^{-1}, \text{ where conversion factor} \\ 0.38 Z = 0.5 \times 0.224170 \text{ converts from} \\ \text{kg C m}^{-2} \text{ to T DM ac}^{-1}. \end{array} $	$G_{ag} \times \begin{cases} \widehat{FC}_{d,100h,max} - 0.85\omega_{o/*} & \omega_{o/*} < 0.38 \\ 1.06 - 1.06\omega_{o/*} & \omega_{o/*} > 0.38 \end{cases}$		
$FC_{d,1000h}$	$G_{ag}(1-\mu_S) \times \times min(0.45, [0.8-0.8 \times \omega_{d,100h}])$	$max\left(0,0.8-0.8\omega_{o/*}\right)$	$(1-\mu_S) \times max \left(0, 0.8 - 0.8 \frac{\omega_{d,1000h}}{0.544}\right)$	$1 - \left(\frac{6.6 - X}{6.6}\right)^2$, where $X = \begin{cases} 1.6107058 - (1.4756 \times \omega_{duff}) & \omega_{duff} \leq 0.\\ 0.5579 \times exp(-3[\omega_{duff} - 0.7]) & \omega_{duff} > 0. \end{cases}$	$G_{ag} \times \left(\widehat{FC}_{d,1000h,max} - 0.8 \times \omega_{o/*}\right)$ 7		
$FC_{l,1h}$	$G_{ag} \times FA_{w,Cr}$	CF	See LPJ-GUESS-SPITFIRE	n/a	$G_{ag} \times FA_{w,Ca+Cr} \times CF$		
$FC_{l,10h}$	$G_{ag} \times FA_{w,Cr} \times 0.05$	0.05CF	See LPJ-GUESS-SPITFIRE	n/a	$G_{ag} \times FA_{w,Ca+Cr} \times CF$		
$FC_{l,100h}$	0	See JSBACH-SPITFIRE	See JSBACH-SPITFIRE	n/a	$G_{ag} \times FA_{w,Ca+Cr} \times CF$		
$FC_{l,1000h}$	0	See JSBACH-SPITFIRE	See JSBACH-SPITFIRE	n/a	$G_{ag} \times FA_{w,Ca+Cr} \times 0.05CF$		
$FC_{l,leaf}$	n/a	n/a	n/a	0.98 if crown fire $(FA_{w,Cr} = 1)$, 0 otherwise (woody PFTs only).	n/a		
$FC_{l,grass}$	$\begin{cases} 1.0 & \omega_{l,g/*} < 0.18 \\ 1.11 - 0.62\omega_{l,g/*} & 0.18 \le \omega_{l,g/*} \le 0.73 \\ 2.45 - 2.45\omega_{l,g/*} & \omega_{l,g/*} > 0.73 \end{cases}$	See JSBACH-SPITFIRE	$G_{ag} imes FC_{d,1h}$	0.9	See JSBACH-SPITFIRE		
$FC_{l,stem}$	n/a	n/a	n/a	Fine branches: 0.98 if crown fire $(FA_{w,Cr}=1)$, 0 otherwise. Other wood, roots: 0.	n/a		
MC-Fi	MC2 (Bachelet et al., Yes 2015; Sheehan et al., 2015)	Yes Yes			nique Bachelet nique@consbio.org)		
	IIDEE-SPITFIRE ORCHIDEE Yes t al., 2014, 2015)	Yes Yes, but not in FireMIP	No 13 2	2 0.5°, daily Chao (chao.	Yue yue@lsce.ipsl.fr)		

Rabin et al. (2017, Geosci. Mod. Dev.)

Burned area trends

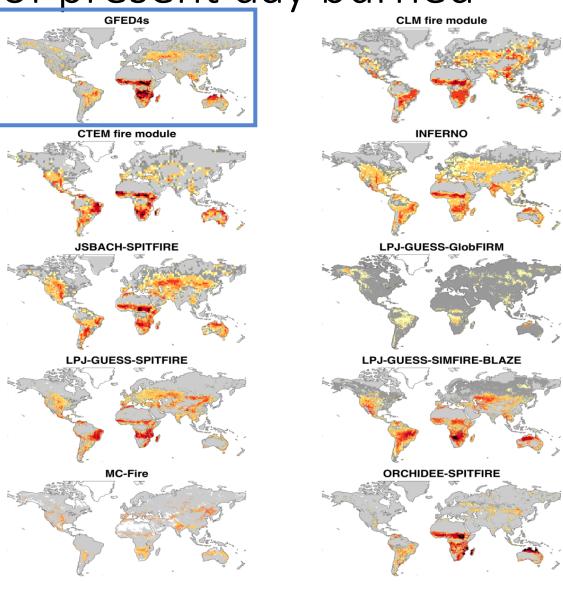


Spatial pattern of present day burned

area

 Most: Too much BA in tropics, not enough elsewhere

- Too much BA in Med. Basin, W. USA
- Too much BA in S.
 America, too little in
 Africa & Australia



Burnt fraction (%)

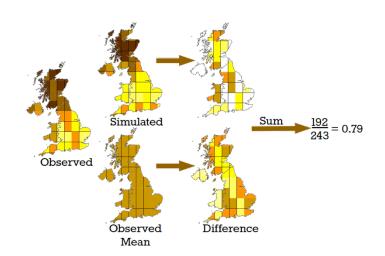
100%

Benchmarking (Kelley et al., 2013)

- Quantify how well a model performs across a comprehensive range of important processes.
- Take advantage of the large number of global datasets to evaluate key processes in DGVMs.
- Allows identification of model weaknesses and processes that require improvement.
- Allows model inter-comparison.
 - Using specifically designed metrics to quantify differences between model and observation (Kelley et al., 2013, BG).

Normalized mean error:

NME =
$$\sum_{i} |y_{i} - x_{i}| / \sum_{i} |x_{i} - \bar{x}|$$



Benchmarking results

	burnt area			fire C emission	fire seasonality	1	carbon in vegetation
	GFED4s	GFED4	Inter-annual variability	GFAS	concentration	phase	Avitabile et al. 2016
mean	1	1	1	1	1		1
random	1.18	1.14	1.3	1.21	1.33	0.43	1.33
CLM	0.64	0.69	0.96	0.82	1.4	0.44	0.72
CTEM	0.78	0.85	1.27	0.94	1.16	0.44	0.9
JULES-INFERNO	0.69	0.77	1.21	0.78	1.44	0.47	0.89
JSBACH-SPITFIRE	0.78	0.89	0.86	1.00	1.32	0.43	
LPJ-GUESS-GlobFIRM	0.75	0.71	1.05	1.72			0.81
LPJ-GUESS-SPITFIRE	0.84	0.98	0.89	0.95	1.07	0.39	0.72
LPJ-GUESS-SIMFIRE-BLAZE	0.83	0.98	1.29	0.97	1.34	0.43	0.71
MC2	0.8	0.79	0.88	0.96			0.85
ORCHIDEE-SPITFIRE	0.69	0.82	1.45	1.18	1.05	0.38	0.73

Spinoff papers

- Patterns: Emissions, seasonality
- Sensitivity runs: Drivers of fire, diagnosing model behavior
- Regional analyses: Boreal, Mediterranean
- Fire impacts: Vegetation, fuel consumption, water cycling

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- FireMIP collaborators

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 - EU FP7 projects BACCHUS (grant agreement no. 603445)
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